Exploratory Analysis of Rainfall Data in India for Agriculture Machine Learning

Submitted by

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1. INTRODUCTION:

1.1. Project Overview:

Predicting weather events to stop loss of life to humans and the environment due to changing climatic conditions and irregular weather patterns is a critical issue facing humanity. Since there have been significant climatic changes in recent years, appropriate preventive actions are required. Floods may occur as a result of heavy rain. Flash floods can be disastrous. Disasters are occurring more frequently, more intensely, and with greater magnitude due to climate change. As a result, more According to the UN Office for Disaster Risk Reduction, weather-related phenomena like heatwaves, storms, floods, and droughts have been responsible for 90% of severe disasters during the past 20 years (UNISDR). Natural disasters are happening more frequently and with greater force. Predictions and emergency preparedness are challenging because of changing weather patterns. Thus, by conducting an exploratory study of the data gathered, we concentrate on the accurate estimation of the likelihood that a flood would occur in a certain place and the recommendation of a nearby evacuation area.

1.2. Purpose

to plan a disaster management system by predicting a flood event in order to reduce flood risk by suggesting an evacuation location from flood hazard zones, which eventually aids in managing the environment and water resource system. By building a model and choosing the most accurate prediction algorithm from the available classifiers, this also serves an early warning system function. The occurrence of flash floods has the potential to be extremely harmful to society. People who live close to riverbeds are first and foremost affected, providing them with the shelter they required and removing them from the danger zones. It has been challenging to anticipate the occurrence of floods using conventional approaches, which has resulted in enormous destruction, due to the irregular change in climate patterns. Consequently, new approaches are developed to deal with severe situations and deal with flash floods.

In order to lessen the damage that a flash flood would cause, technology needs to be more thoughtful. It is now considerably simpler to anticipate floods and provide safe havens for evacuees in the current day. The rainfall readings are collected, combined from many resources, curated, mined, evaluated, and predictions are made over patterns in hazardous locations that are vulnerable to destruction and devastation. The prognosis is accompanied with a list of society-related recommendations. Early warning systems are included into climate change adaption strategies.

2. LITERATURE SURVEY:

2.1. Existing Solutions

In this research, the concept of flood prediction utilising artificial neural networks (ANN) and the Internet of Things is discussed. This system monitors the river water level, humidity, temperature, pressure, and rainfall on a regular basis to provide temporal correlative data for flood prediction analysis. The nature of flood data is dynamic and non-linear. The system is informed by the sensors, which read the data. With such figures, the projection is complete, and a choice is made regarding the likelihood of a flood.

2.2. References

- Xiaobo Zhang;Sachi Nandan Mohanty;Ajaya Kumar Parida;Subhendu Kumar Pani;Bin Dong;Xiaochun Cheng.
 Annual and Non-Monsoon Rainfall Prediction Modelling Using SVR-MLP: An Empirical Study From Odisha.
 (IEEE)(2020)
- Shilpa Manandhar; Soumyabrata Dev; Yee Hui Lee; Yu Song Meng; Stefan Winkler A Data-Driven Approach for Accurate Rainfall Prediction. (IEEE)(2019)
- 3. Nana Kofi Ahoi Appiah-Badu; Yaw Marfo Missah; Leonard K. Amekudzi; Najim Ussiph; Twum Frimpong; Emmanuel Ahene Rainfall Prediction Using Machine Learning Algorithms for the Various Ecological Zones of Ghana. (IEEE) (2021)
- 4. Deepali Patil Shree L.R.Rainfall Prediction using Linear approach & Neural Networks and Crop

 Recommendation based on DecisionTree.(IEEE)(2020)
- A. Haidar, B. Verma. Monthly Rainfall Forecasting Using One-Dimensional Deep Convolutional Neural Network (IEEE)

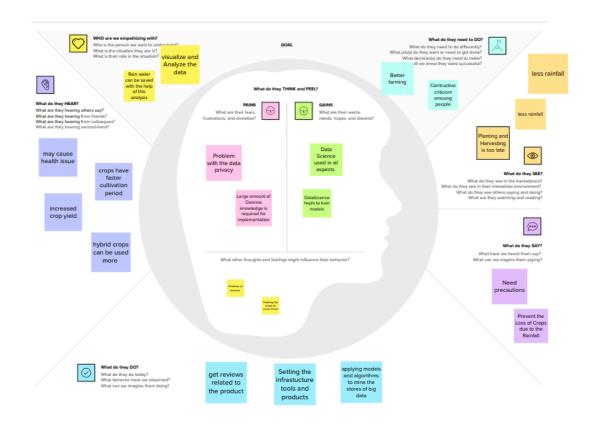
2.3. Problem Statement Definition

Rainfall has been a major concern these days. Irregular heavy rainfall may lead to the destruction of crops, heavy floods that can cause harm to human life.

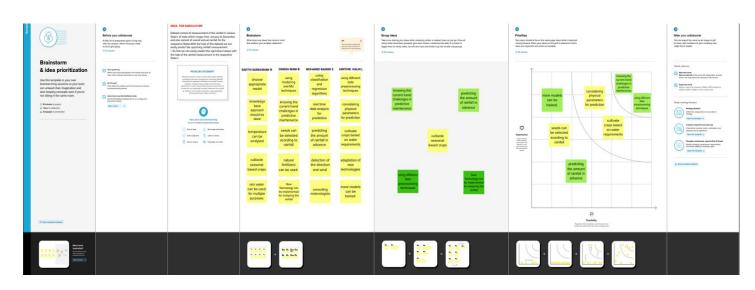
It is important to exactly determine the rainfall for effective use of water resources, crop productivity, and pre-planning of water structures.

3. IDEATION AND PROPOSED SOLUTION:

3.1. Empathy Map Canvas:



3.2. Ideation and Brainstorming:



3.3. Proposed Solution:

- ► The dataset consist of measurement of the rainfall in various State's of India which ranges from January to December and also consist of overall annual rainfall for the respective States
- ► With the help of this dataset we can easily predict the upcoming rainfall measurement
- ► So that we can easily predict the agricultural status with the help of the rainfall measurement in the respective State's

3.4. Solution Fit

Project Title: Exploratory Analysis of RainFall Data in India for Agriculture Project Design Phase-I - Solution Fit Template

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4. REQUIREMENT ANALYSIS:

4.1. Functional Requirements

S.No	Component	Description	Technology	
1.	User Interface	The user interacts with the application through a web UI and a chatbot	HTML, CSS, python, Flask	
2.	Application Logic-1	Logic for registration Registration	Python	
3.	Application Logic-2	Logic for login to the application	Python	
4.	Application Logic-3	Integrating machine learning model and the webpage	Flask	
5.	Database	Numeric data	MySQL	
6.	File Storage	To store files such as prediction report	Local Filesystem	
7.	External API	Allows developers access to critical forecasts, alerts, and observations, along with other weather data.	IBM Weather API	

8.	Machine Learning Model	Predictive modeling is a statistical technique using machine learning and data mining to predict and forecast likely future outcomes with the aid of historical and existing data	Predictive modeling
9.	Infrastructure (Server)	Application Deployment on Local System Local Server Configuration: built-in flask web server	Flask web server

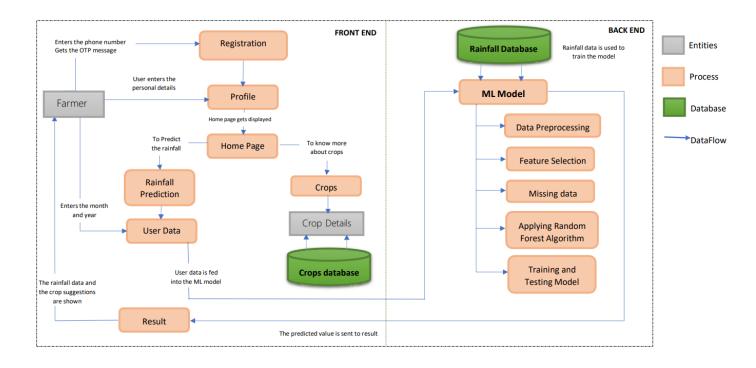
4.2. Non Functional Requirements

S.No	Characteristics	Description	Technology
1.	Open-Source Frameworks	Flask	Micro web framework written in Python
2.	Security Implementations	Basic HTTP authentication, Session based authentication, User Registration, Login Tracking	Flask Security

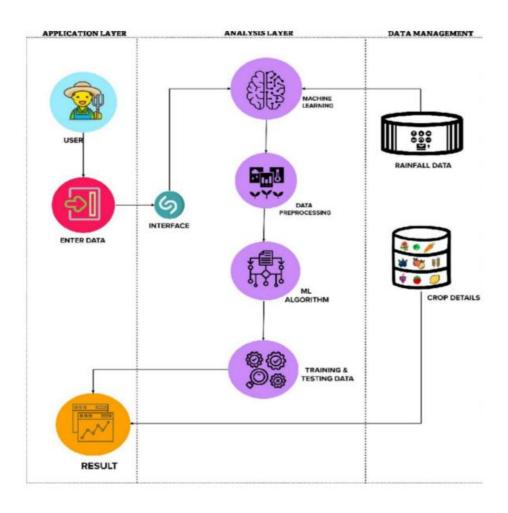
3.	Scalable Architecture	Size is everything, and Flask's status as a microframework means that you can use it to grow a tech project such as a web app incredibly quickly. Its simplicity of use and few dependencies enable it to run smoothly even as it scales up and up.	Flask
4.	Availability	Higher compatibility with latest technologies and allows customization	Flask
5.	Performance	 Integrated support for unit testing. RESTful request dispatching. Uses Jinja templating. Support for secure cookies (client side sessions) 100% WSGI 1.0 compliant. 	Flask

5. PROJECT DESIGN

5.1. Data Flow Diagram:



5.2. Solution And Technical Architecture:



5.3. User Stories:

SPRINT NO	FUNCTION PERFORMED	TASK	STORY POINTS	PRIORITY	TEAM MEMBERS
SPRINT-1	Data Collection and processing	The needed data for processing and cleaning of the same has been done	20	High	L.Hirithik kalin, B.Dineshmani, S.Mohammed Baseer, N.Sakthi Saravanan
SPRINT-2	Model Building	Splitting of data into two sets and testing of the data is done	20	High	L.Hirithik kalin, B.Dineshmani, S.Mohammed Baseer, N.Sakthi Saravanan
SPRINT-3	Integrating with flask	Python has been integrated with flask framework	20	High	L.Hirithik kalin, B.Dineshmani, S.Mohammed Baseer, N.Sakthi Saravanan
SPRINT-4	Deployment of code	Deployment of code has been done	20	High	L.Hirithik kalin, B.Dineshmani, S.Mohammed Baseer, N.Sakthi Saravanan

6. PROJECT PLANNING & SCHEDULING:

6.1. Planning & Estimation:

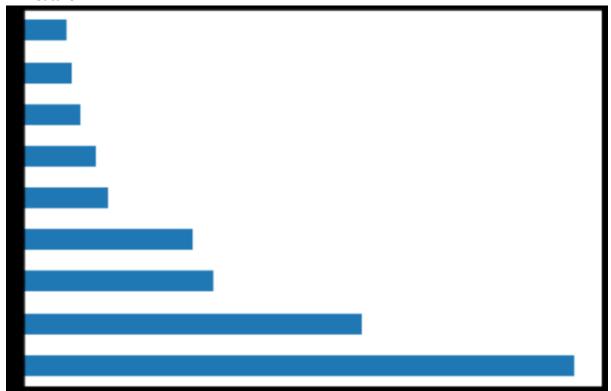
Sprint	Tot al Stor y Poin ts	Duration	Sprint StartDate	Sprint End Date (Planned)	Story Points Completed (ason Planned End Date)	Sprint Release Date (Actual)
Sprint-1	20	6 Days	24 Oct 2022	29 Oct 2022	2 0	29 Oct 2022
Sprint-2	20	6 Days	31 Oct 2022	05 Nov 2022	2 0	05 Nov 2022
Sprin t-3	20	6 Days	07 Nov 2022	12 Nov 2022	2 0	12 Nov 2022
Sprint-4	20	6 Days	14 Nov 2022	19 Nov 2022	2 0	19 Nov 2022

7. CODING & SOLUTIONING

7.1 Feature 1

1	0.015	0.18	0.25	0.35	0.53	0.61	0.73	0.62	0.62	0.59	0.53	0.3	0.23	0.84
0.015		0.053	0.029	0.022	0.0093	0.0071	0.011	-0.010	0005	0.004	D 002	20.023	0.011	0.011
0.18	0.053	1	0.45	0.39	0.21	0.13	0.036	-0.053	0.013	0.023	0.01	0.068	0.23	0.11
0.25	-0.029	0.45	1	0.58	0.36		0.028	0.014	0.071	0.079	0.007	60.021	0.13	0.18
0.35	0.022	0.39	0.58	1	0.56	0.36		0.096	0 14		0.087	0 0091	0.13	0.32
0.53	0.0093	0.21	0.36	0.56	1	0.65	0.45	0.26	0.25	0.38	0.37	0.17		0.58
0.61	0.0071	0.13		0.36	0.65	1	0.56	0.33	0.33	0.49	0.53	0.35	0.25	0.7
0.73	-0.011	0.036	0.028	017	0.45	0.56	1	0.74	0.65	0.55	0.49	0.23	0.087	0.89
0.62	-0.017	40.053	0.014	0.096	0.26	0.33	0.74		0.69	0.51	0.3	0.041	-0.022	0.81
0.62	.0005	D.013	0.071	0.14	0.25	0.33	0.65	0.69	1	0.5	0.25	0.019	.0004	0.76
0.59	0.004	70.023	0.079		0.38	0.49	0.55	0.51	0.5	1	0.39	0.15	0.12	0.72
0.53	0.0022	0.01	0.007	60 087	0.37	0.53	0.49	0.3	0.25	0.39	1	0.48	0.28	0.59
0.3	0.023	0.068	-0.021	0.0091	0.17	0.35	0.23	0.041	0.019	0.15	0.48	1	0.45	0.31
0.23	0.011	0.23	0.13	0.13	0.13	0.25	0.087	-0.020	0004	40.12	0.28	0.45	1	0.21
0.84	-0.011	0 11		0.32	0.58	0.7	0.89	0.81	0.76	0.72	0.59	0.31	0.21	1

7.1 Feature 2



10. ADVANTAGES & DISADVANTAGES

Advantage

- 1. Farmers can schedule the planting and harvesting of their crops.
- 2. When planning a vacation, people can choose a location and a time that will maximise the favourable weather.
- 3. Surfers are aware of when big waves are predicted.
- 4. If hurricanes or floods are predicted, certain areas can be evacuated.
- 5. Accurate weather predictions are extremely important for shipping and aircraft.

Disadvantage

- 1. Weather is exceedingly challenging to predict accurately.
- 2. Monitoring so many different factors from so many different sources is expensive.
- 3. The cost of the computers required to do the requisite millions of calculations is high.
- 4. If the weather does not match the forecast, the weather forecasters are held responsible.

11. CONCLUSION

One of the most important things these days is the ability to predict the weather. To make risk management systems better and to automatically and scientifically predict the weather for the upcoming days. In order to help with weather prediction, numerous models have been developing. Using six different machine learning (ML) algorithms—Cat Boost Classifier, Random For set Classifier, Logistic Regression, Gaussian NB, KNN, and XGB Classifier—we built a weather prediction web application from scratch in this study. The findings from all six models are listed in the results section, together with their accuracy, error rate, mean absolute error, root mean squared error, relative squared error, root relative squared error, and modelling time. The results demonstrate that, in comparison to all other classifiers utilised, the Cat Boost Classifier and the XGB Classifier have produced results with a high degree of accuracy. The Cat Boost Classifier surpasses all other classifiers in tackling the problem at hand while taking the least amount of time to create the model.

12. FUTURE SCOPE

The WEATHER FORECASTING programme will get new features in next releases, like:

- 1. Live Location Tracking News on Current Disasters Week's Worth of Weather deploy as an Android application
- 2. Assistance in determining which crop will be most suitable given the weather